Econometrics: What's It All About, Alfie?

Using sample data on observable variables to learn about economic relationships, the functional relationships among economic variables.

Econometrics consists mainly of:

- estimating economic relationships from sample data
- *testing hypotheses* about how economic variables are related
 - the *existence* of relationships between variables
 - the *direction* of the relationships between one economic variable -- the dependent or outcome variable -- and its hypothesized observable determinants
 - the *magnitude* of the relationships between a dependent variable and the independent variables that are thought to determine it.

Sample data consist of *observations* on *randomly selected* members of populations of economic agents (individual persons, households or families, firms) or other units of observation (industries, provinces or states, countries).

Example 1

We wish to investigate empirically the determinants of households' food expenditures, in particular the relationship between households' food expenditures and households' incomes.

Sample data consist of **a random sample of 38 households** from the population of all households. For each household in the random sample, we have observations on three observable variables:

- foodexp = annual food expenditure of household, thousands of dollars per year
- income = annual income of household, thousands of dollars per year
- hhsize = household size, number of persons in household

. list foodexp income hhsize

26.10.88261.041227.18.56182.4691	28. 11.629 44.208 2		29.18.06749.467530.14.53925.905531.19.19279.178532.25.91875.811333.28.83382.718634.15.86948.311435.14.9142.4945	27. 28.	18.561 11.629	82.469 44.208	1
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. describe

Contains data from foodexp.dta
 obs:
 38

 vars:
 3

 size:
 608 (99.9% of memory free)
7 Sep 2000 23:30 vars: size: food expenditure, thousands \$ 1. foodexp float %9.0g per yr 2. income float %9.0g household income, thousands \$ per yr 3. hhsize float %9.0g household size, persons per hh -----_____ Sorted by: Note: dataset has changed since last saved

. summarize

Variable	Obs	Mean	Std. Dev.	Min	Max
foodexp income hhsize	38 38 38 38	15.95282 58.44434 3.578947	5.624341 19.93156 1.702646	7.431 25.905 1	28.98 88.233 7

<u>Question</u>: What relationship generated these sample data? What is the data generating process?

<u>Answer</u>: We postulate that each population value of foodexp, denoted as foodexp_i, is generated by a relationship of the form:

 $foodexp_i = f(income_i, hhsize_i) + u_i \iff the$ *population*regression equation

where

- $income_i = on$ *independent* or *explanatory* variable that we think might explain the dependent variable food exp_i
 - = the annual income of household i (thousands of \$ per year)
- hhsize_i = a second *independent* or *explanatory* variable that we think might explain the dependent variable food exp_i
 - = household size, measured by the number of persons in the household
- f(income_i, hhsize_i) = a **population regression function** representing the systematic relationship of food exp_i to the independent or explanatory variables income_i and hhsize_i;
- u_i = an *unobservable random* error term representing all *unknown* and *unmeasured* variables that determine the individual population values of food exp_i

<u>**Question:**</u> What mathematical form does the population regression function $f(\text{income}_i, \text{hhsize}_i)$ take?

<u>Answer</u>: We hypothesize that the **population regression function** -- or **PRF** -- is a linear function:

 $f(income_i, hhsize_i) = \beta_0 + \beta_1 income_i + \beta_2 hhsize_i$

Implication: The population regression equation -- the PRE -- is therefore

food exp_i = f(income_i, hhsize_i) + u_i = $\beta_0 + \beta_1$ income_i + β_2 hhsize_i + u_i

• Observable Variables:

 $foodexp_i \equiv the value of the dependent variable foodexp for the i-th household income_i \equiv the value of the independent variable income for the i-th household hhsize_i \equiv the value of the independent variable hhsize for the i-th household$

• Unobservable Variable:

 $u_i \equiv$ the value of the random error term for the i-th household in the population

• Unknown Parameters: the regression coefficients β_0 , β_1 and β_2

- β_0 = the *intercept* coefficient
- β_1 = the *slope* coefficient on income_i
- β_2 = the *slope* coefficient on hhsize_i

The population values of the regression coefficients β_0 , β_1 and β_2 are *unknown*.

Example 2

We wish to investigate empirically the determinants of paid workers' wage rates. In particular, we want to investigate whether male and female workers with the same characteristics on average earn the same wage rate.

Sample data consist of **a random sample of 526 paid workers** from the 1976 US population of all paid workers in the employed labour force.

For each paid worker in the random sample, we have observations on six observable variables:

wage	= average hourly earnings of paid worker, dollars per hour
ed	= years of education completed by paid worker, years
exp	= years of potential work experience of paid worker, years
ten	= tenure, or years with current employer, of paid worker, years
female	= 1 if paid worker is female, = 0 otherwise
married	= 1 if paid worker is married, = 0 otherwise

. describe

Contains data obs: vars: size:	526 6	l.dta).7% of memory free)	16 Apr 2000 16:18
1. wage 2. ed 3. exp	float % float % float %	\$9.0g	average hourly earnings, \$/hour years of education years of potential work experience
4. ten	float %	}9.0g	tenure = years with current employer
5. female 6. married		5	=1 if female, =0 otherwise =1 if married, =0 otherwise

. list wage ed exp ten female married

1. 2. 3. 4. 5. 6. 7. 8. 9. 10. 11. 12. 13. 14. 15. 16. 17. 18. 19. 20.	wage 21.86 5.5 3.75 10 3.5 6.67 3.88 5.91 5.9 10 4.55 10 6 5.43 2.83 6.8 6.76 4.51	ed 12 12 2 12 13 12 12 12 12 12 12 17 16 8 13 9 12 14 10 12 12 12	exp 24 18 39 31 1 35 12 14 14 5 34 9 8 31 13 10 1 14 19 5	ten 16 3 13 2 0 10 3 6 7 3 2 0 0 0 9 0 3 0 10 3 2	female 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	married 1 0 1 1 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1
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. summarize wage ed exp ten female married

Variable	Obs	Mean	Std. Dev.	Min	Max
wage	 526	5.896103	3.693086	.53	24.98
ed	526	12.56274	2.769022	0	18
exp	526	17.01711	13.57216	1	51
ten	526	5.104563	7.224462	0	44
female	526	.4790875	.500038	0	1
married	526	.608365	.4885804	0	1

<u>Question</u>: What relationship generated these sample data? What is the data generating process?

<u>Answer</u>: We postulate that **each population value of wage**, denoted as **wage**_i, is generated by a *population* regression equation of the form:

 $wage_i = f(ed_i, exp_i, ten_i, female_i, married_i) + u_i$

where:

- $wage_i = the$ *dependent*or*outcome*variable we are trying to explain
 - = the average hourly earnings of paid worker i (dollars per hour)
- ed_i = one *independent* or *explanatory* variable that we think might explain the dependent variable wage_i
 - = the years of education completed by paid worker i (years)
- exp_i = a second *independent* or *explanatory* variable that might explain wage_i = the potential work experience accumulated by paid worker i (years)
- ten_i = a third *independent* or *explanatory* variable that might explain wage_i = tenure, years with current employer, of paid worker i (years)
- female_i = a fourth *independent* or *explanatory* variable that might affect wage_i = 1 if paid worker i is female, = 0 otherwise
- married_i = a fifth *independent* or *explanatory* variable that we think might explain the dependent variable wage_i
 - = 1 if paid worker i is married, = 0 otherwise

 $f(ed_i, exp_i, ten_i, female_i, married_i)$

- a population regression function representing the systematic relationship of wage_i to the independent variables ed_i, exp_i, ten_i, female_i and married_i
- u_i = an *unobservable random* error term representing all *unknown* variables and *unmeasured* that determine the individual population values of wage_i

<u>Question</u>: What mathematical form does the population regression function, or PRF, $f(ed_i, \dots, married_i)$ take?

<u>Answer</u>: We hypothesize that the **population regression function** -- or **PRF** -- is a *linear* function.

 $f(ed_1, \dots, married_i) = \beta_0 + \beta_1 ed_1 + \beta_2 exp_1 + \beta_3 ten_1 + \beta_4 female_1 + \beta_5 married_1$

Implication: The population regression equation -- the PRE -- is therefore

wage_i = f(ed_i, exp_i, ten_i, female_i, married_i) + u_i
=
$$\beta_0 + \beta_1 ed_i + \beta_2 exp_i + \beta_3 ten_i + \beta_4 female_i + \beta_5 married_i + u_i$$

• Observable Variables:

wage_i \equiv the value of the dependent variable wage for the i-th employee ed_i \equiv the value of the independent variable ed for the i-th employee exp_i \equiv the value of the independent variable exp for the i-th employee ten_i \equiv the value of the independent variable ten for the i-th employee female_i \equiv the value of the independent variable female for the i-th employee married_i \equiv the value of the independent variable married for the i-th employee

• Unobservable Variable:

 $u_i \equiv$ the value of the random error term for the i-th paid worker in the population

- Unknown Parameters: the regression coefficients β_0 , β_1 , β_2 , β_3 , β_4 and β_5
 - β_0 = the *intercept* coefficient
 - β_1 = the *slope* coefficient on ed_i
 - β_2 = the *slope* coefficient on exp_i
 - β_3 = the *slope* coefficient on ten_i
 - β_4 = the *slope* coefficient on female_i
 - β_5 = the *slope* coefficient on married_i

<u>Our task</u>: To learn how to compute from sample data reliable estimates of the regression coefficients β_0 , β_1 , β_2 , β_3 , β_4 and β_5 .

The Four Elements of Econometrics

<u>Data</u>

Collecting and coding the sample data, the raw material of econometrics.

Most economic data is *observational*, or *non-experimental*, data (as distinct from *experimental* data generated under controlled experimental conditions).

Specification

Specification of the *econometric model* that we think (hope) generated the sample data -- that is, specification of the **data generating process** (or DGP).

An *econometric model* consists of <u>two</u> components:

- 1. An *economic model*: specifies the *dependent* or *outcome* variable to be explained and the *independent* or *explanatory* variables that we think are related to the dependent variable of interest.
 - Often suggested or derived from economic theory.
 - Sometimes obtained from informal intuition and observation.
- 2. A *statistical model*: specifies the statistical elements of the relationship under investigation, in particular the *statistical properties* of the *random* variables in the relationship.

Estimation

Consists of **using the assembled** *sample data* on the *observable* **variables** in the model **to compute** *estimates* of the **numerical values** of all the **unknown parameters** in the model.

Inference

Consists of **using the parameter estimates** computed from sample data **to test hypotheses** about the **numerical values** of the **unknown** *population* **parameters** that describe the behaviour of the population from which the sample was selected.

Scientific Method

The collection of principles and processes necessary for scientific investigation, including:

- 1. rules for concept formation
- 2. rules for conducting observations and experiments
- 3. rules for validating hypotheses by observations or experiments

Econometrics is that branch of economics -- the dismal science -- which is concerned with items 2 and 3 in the above list.

Recap

We have considered **two examples** of what are generically called *linear regression equations* or *linear regression models*.

Example 1 -- a linear regression model for household food expenditure:

food $exp_i = \beta_0 + \beta_1 income_i + \beta_2 hhsize_i + u_i$

Example 2 -- a linear regression model for paid workers' wage rates:

 $wage_{i} = \beta_{0} + \beta_{1}ed_{i} + \beta_{2} exp_{i} + \beta_{3}ten_{i} + \beta_{4}female_{i} + \beta_{5}married_{i} + u_{i}$

Regression analysis has two fundamental tasks:

- 1. <u>Estimation</u>: computing from *sample* data reliable *estimates* of the *numerical* values of the regression coefficients β_j (j = 0, 1, ..., K), and hence of the population regression function.
- Inference: using sample estimates of the regression coefficients β_j (j = 0, 1, ..., K) to test hypotheses about the *population* values of the unknown regression coefficients -- i.e., to *infer* from sample estimates the true *population* values of the regression coefficients within specified margins of statistical error.